



# On the path structure of a semimartingale arising from monotone probability theory

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**Abstract.** Let  $X$  be the unique normal martingale such that  $X_0 = 0$  and

$$d[X]_t = (1 - t - X_{t-}) dX_t + dt$$

and let  $Y_t := X_t + t$  for all  $t \geq 0$ ; the semimartingale  $Y$  arises in quantum probability, where it is the monotone-independent analogue of the Poisson process. The trajectories of  $Y$  are examined and various probabilistic properties are derived; in particular, the level set  $\{t \geq 0: Y_t = 1\}$  is shown to be non-empty, compact, perfect and of zero Lebesgue measure. The local times of  $Y$  are found to be trivial except for that at level 1; consequently, the jumps of  $Y$  are not locally summable.

**Résumé.** Soit  $X$  l'unique martingale normale telle que  $X_0 = 0$  et

$$d[X]_t = (1 - t - X_{t-}) dX_t + dt$$

et soit  $Y_t := X_t + t$  pour tout  $t \geq 0$ ; la semimartingale  $Y$  se manifeste dans la théorie des probabilités quantiques, où c'est analogue du processus de Poisson pour l'indépendance monotone. Les trajectoires de  $Y$  sont examinées et diverses propriétés probabilistes sont déduites; en particulier, l'ensemble de niveau  $\{t \geq 0: Y_t = 1\}$  est montré être non vide, compact, parfait et de mesure de Lebesgue nulle. Les temps locaux de  $Y$  sont trouvés être triviaux sauf celui au niveau 1; par conséquent les sauts de  $Y$  ne sont pas localement sommables.

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## 0. Introduction

The first Azéma martingale, that is, the unique (in law) normal martingale  $M$  such that  $M_0 = 0$  and

$$d[M]_t = -M_{t-} dM_t + dt,$$

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has been the subject of much interest since its appearance in [3], Proposition 118 (see, for example, [4, 13] and [17], Section IV.6); it was the first example to be found of a process without independent increments which possesses the chaotic-representation property. It shall henceforth be referred to as *Azéma's martingale*.

From a quantum-stochastic viewpoint, the process  $M$  may be obtained by applying Attal's D transform ([1], Section IV) to the Wiener process. Furthermore, thanks to the factorisation of D provided by vacuum-adapted calculus [5],  $M$  appears as a natural object in monotone-independent probability theory; the distribution of  $M_t$  (the arcsine law) is a central-limit law which plays a rôle analogous to that played by the Gaussian distribution in the classical framework ([16], Theorem 3.1).

The Poisson distribution also occurs as a limit (the *law of small numbers*): if, for all  $n \geq 1$ ,  $(x_{n,m})_{m=1}^n$  is a collection of independent, identically distributed random variables and there exists a constant  $\lambda > 0$  such that

$$\lim_{n \rightarrow \infty} n\mathbb{E}[x_{n,1}^k] = \lambda \quad \forall k \geq 1,$$

then  $x_{n,1} + \dots + x_{n,n}$  converges in distribution to the Poisson law with mean  $\lambda$ . (A simple proof of this result is provided in Appendix A.) In the case where  $x_{n,1}, \dots, x_{n,n}$  are Bernoulli random variables taking the values 0 and 1 with mean  $\lambda/n$ , this is simply the Poisson approximation to the binomial distribution ([8], Example 25.2).

A corresponding theorem holds in the monotone set-up ([16], Theorem 4.1), but now the limit distribution is related to the D transform of the standard Poisson process (with intensity 1 and unit jumps) in the same way as the arcsine law and Azéma's martingale are related above [6]. (This result also holds for free probability: see [20], Theorem 4.) The classical process  $Y$  which results is such that  $Y_t = X_t + t$  for all  $t \geq 0$ , where  $X$  is the unique normal martingale such that  $X_0 = 0$  and

$$d[X]_t = (1 - t - X_{t-}) dX_t + dt.$$

This article extends the sample-path analysis of  $Y$  (and so  $X$ ) which was begun in [7]. Many similarities are found between  $Y$  and Azéma's martingale  $M$ ; for example, they are both determined by a random perfect subset of  $\mathbb{R}_+$  and a collection of binary choices, one for each interval in that subset's complement. In Section 1 some results from the theory of martingales are recalled; Section 2 defines the processes  $X$  and  $Y$  and presents their Markov generators. A random time  $G_\infty$  after which  $Y$  is deterministic is discussed in Section 3: by Proposition 3.1 and Corollary 3.5,  $G_\infty < \infty$  almost surely and, in this case,

$$Y_{t+G_\infty} = -W_{-1}(-\exp(-1-t)) \quad \forall t \geq 0,$$

where  $W_{-1}$  is a certain branch of the inverse to the function  $z \mapsto ze^z$  (see Notation below). In Section 4 the process  $X$  is decomposed into an initial waiting time  $S_0$  which is exponentially distributed and an independent normal martingale  $Z$  which satisfies the same structure equation as  $X$  but has the initial condition  $Z_0 = 1$ ; Lemma 4.2 implies that, for all  $t \geq 0$ ,

$$X_t = \begin{cases} -t & \text{if } t \in [0, S_0[, \\ Z_{t-S_0} - S_0 & \text{if } t \in [S_0, \infty[. \end{cases}$$

Explicit formulae are found for the distribution functions of  $G_\infty$  and  $J$ , a random variable analogous to  $G_\infty$  but for  $Z$  rather than  $X$ . In Section 5 it is shown that  $(H_t := 1 - (Z_t + t)^{-1})_{t \geq 0}$  is a martingale which is related to Azéma's martingale  $M$  by a time change; this gives a simple way to find various properties of the level set  $\mathcal{U} := \{t \geq 0: Y_t = 1\}$  in Section 6. Finally, Section 7 presents some results on the local times of  $Y$ . The appendices contain various supplementary results which are not appropriate for the main text.

### 0.1. Conventions

The underlying probability space is denoted  $(\Omega, \mathcal{F}, \mathbb{P})$  and is assumed to contain a filtration  $(\mathcal{F}_t)_{t \geq 0}$  which generates the  $\sigma$ -algebra  $\mathcal{F}$ . This filtration is supposed to *satisfy the usual conditions*: it is right continuous

and the initial  $\sigma$ -algebra  $\mathcal{F}_0$  contains all the  $\mathbb{P}$ -null sets. Each semimartingale which is considered below has *càdlàg* paths (that is, they are right-continuous with left limits) and two processes  $(X_t)_{t \geq 0}$  and  $(Y_t)_{t \geq 0}$  are taken equal if they are *indistinguishable*:  $\mathbb{P}(X_t = Y_t \text{ for all } t \geq 0) = 1$ . Any quadratic variation or stochastic integral has value 0 at time 0.

## 0.2. Notation

The expression  $\mathbb{1}_P$  is equal to 1 if the proposition  $P$  is true and equal to 0 otherwise; the indicator function of a set  $A$  is denoted by  $\mathbb{1}_A$ . The set of natural numbers is denoted by  $\mathbb{N} := \{1, 2, 3, \dots\}$ , the set of non-negative rational numbers is denoted by  $\mathbb{Q}_+$  and the set of non-negative real numbers is denoted by  $\mathbb{R}_+$ . The branches of the Lambert  $W$  function (that is, the multi-valued inverse to the map  $z \mapsto ze^z$ ) which take (some) real values are denoted by  $W_0$  and  $W_{-1}$ , following the conventions of Corless et al. [10]:

$$W_0(0) = 0, \quad W_0(x) \in [-1, 0[ \quad \text{and} \quad W_{-1}(x) \in ]-\infty, -1] \quad \forall x \in [-e^{-1}, 0[.$$

If  $\Xi$  is a topological space then  $\mathcal{B}(\Xi)$  denotes the Borel  $\sigma$ -algebra on  $\Xi$ . The integral of the process  $X$  by the semimartingale  $R$  will be denoted by  $\int X_t dR_t$  or  $X \cdot R$ , as convenient; the differential notation  $X_t dR_t$  will also be employed. The process  $X$  stopped at  $T$  is denoted by  $X^T$ , that is,  $X_t^T := X_{t \wedge T}$  for all  $t \geq 0$ , where  $x \wedge y$  denotes the minimum of  $x$  and  $y$ . For all  $x$ , the positive part  $x^+ := \max\{x, 0\}$ , the maximum of  $x$  and 0.

## 1. Normal sigma-martingales and time changes

**Remark 1.1.** Let  $A \in \mathcal{F}$  be such that  $\mathbb{P}(A) > 0$ . If  $\mathcal{G}$  is a sub- $\sigma$ -algebra of  $\mathcal{F}$  such that  $A \in \mathcal{G}$  then

$$\tilde{\mathcal{G}} := \{B \subseteq \Omega: B \cap A \in \mathcal{G}\}$$

is a  $\sigma$ -algebra containing  $\mathcal{G}$ ; the map  $\mathcal{G} \mapsto \tilde{\mathcal{G}}$  preserves inclusions and arbitrary intersections. If

$$\tilde{\mathbb{P}} := \mathbb{P}(\cdot | A) : \tilde{\mathcal{F}} \rightarrow [0, 1]; \quad B \mapsto \frac{\mathbb{P}(B \cap A)}{\mathbb{P}(A)},$$

then  $(\Omega, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$  is a complete probability space; if  $(\mathcal{G})_{t \geq 0}$  is a filtration in  $(\Omega, \mathcal{F}, \mathbb{P})$  satisfying the usual conditions then  $(\tilde{\mathcal{G}}_t)_{t \geq 0}$  is a filtration in  $(\Omega, \tilde{\mathcal{G}}, \tilde{\mathbb{P}})$  which satisfies them as well.

If  $T$  is a stopping time for the filtration  $(\mathcal{G}_t)_{t \geq 0}$  then it is also one for  $(\tilde{\mathcal{G}}_t)_{t \geq 0}$  and, if  $B \subseteq \Omega$ ,

$$\begin{aligned} B \in \tilde{\mathcal{G}}_T &\iff B \cap A \in \mathcal{G}_T \iff B \cap A \cap \{T \leq t\} \in \mathcal{G}_t \quad \forall t \geq 0, \\ &\iff B \cap \{T \leq t\} \in \tilde{\mathcal{G}}_t \quad \forall t \geq 0 \iff B \in (\tilde{\mathcal{G}})_T, \end{aligned}$$

so the notation  $\tilde{\mathcal{G}}_T$  is unambiguous.

**Lemma 1.2.** If  $T$  is a stopping time such that  $\mathbb{P}(T < \infty) > 0$  and  $M$  is a local martingale then  $N : t \mapsto \mathbb{1}_{T < \infty}(M_{t+T} - M_T)$  is a local martingale for the conditional probability measure  $\tilde{\mathbb{P}} := \mathbb{P}(\cdot | T < \infty)$  and the filtration  $(\tilde{\mathcal{F}}_{t+T})_{t \geq 0}$ , such that

$$[N]_t = \mathbb{1}_{T < \infty}([M]_{t+T} - [M]_T) \quad \forall t \geq 0.$$

**Proof.** If  $T < \infty$  almost surely and  $M$  is uniformly integrable then the first part is immediate, by optional sampling ([18], Theorem II.77.5), and holds in general by localisation and conditioning. The second claim may be verified by realising  $[N]$  as a limit of sums in the usual manner (see [17], Theorem II.22, for example).  $\square$

**Definition 1.3.** A martingale  $M$  is normal if  $t \mapsto (M_t - M_0)^2 - t$  is also a martingale. (If  $M_0$  is square integrable then this is equivalent to  $t \mapsto M_t^2 - t$  being a martingale, but in general it is a weaker condition.)

**Definition 1.4.** A semimartingale  $M$  is a sigma-martingale if it can be written as  $K \cdot N$ , where  $N$  is a local martingale and  $K$  is a predictable,  $N$ -integrable process. Equivalently, there exists an increasing sequence  $(A_n)_{n \geq 1}$  of predictable sets such that  $\bigcup_{n \geq 1} A_n = \mathbb{R}_+ \times \Omega$  and  $1_{A_n} \cdot M \in H^1$  for all  $n \geq 1$ , where  $H^1$  denotes the Banach space of martingales  $M$  with  $\|M\|_{H^1} := \mathbb{E}[|M|_\infty^{1/2}] < \infty$ . Every local martingale is a sigma-martingale and if  $M$  is a sigma-martingale then so is  $H \cdot M$  for any predictable,  $M$ -integrable process  $H$ . (The class of sigma-martingales, so named by Delbaen and Schachermayer in [11], was introduced by Chou in [9], where it is denoted  $(\Sigma_m)$ ; the equivalence mentioned above is due to Émery ([12], Proposition 2).)

**Theorem 1.5** ([14]). If  $M$  is a semimartingale with  $M_0 = 0$  then the following are equivalent:

- (i)  $M$  and  $t \mapsto M_t^2 - t$  are sigma-martingales;
- (ii)  $M$  and  $t \mapsto [M]_t - t$  are sigma-martingales;
- (iii)  $M$  and  $t \mapsto M_t^2 - t$  are martingales;
- (iv)  $M$  and  $t \mapsto [M]_t - t$  are martingales.

**Proof.** Since  $M^2 - [M] = 2M_- \cdot M$ , the equivalence of (i) and (ii) is immediate; it also follows from this that (iv) implies (iii) ([17], Corollary 3 to Theorem II.27). To complete the proof it suffices to show that (ii) implies (iv).

Suppose (ii) holds and let  $(A_n)_{n \geq 1}$  be an increasing sequence of predictable sets such that  $\bigcup_{n \geq 1} A_n = \mathbb{R}_+ \times \Omega$  and both  $1_{A_n} \cdot M \in H^1$  and  $1_{A_n} \cdot N \in H^1$  for all  $n \geq 1$ , where  $N : t \mapsto [M]_t - t$ . (Note that if  $X \in H^1$  and  $B$  is a predictable set then  $1_B \cdot X \in H^1$ .) Let  $T$  be a bounded stopping time; since  $1_{A_n} \cdot N$  is a martingale,

$$\mathbb{E}[(1_{A_n} \cdot [M])_T] = \mathbb{E}[(1_{A_n} \cdot N)_T] + \mathbb{E}\left[\int_0^T 1_{A_n} \, ds\right] = \mathbb{E}\left[\int_0^T 1_{A_n} \, ds\right] \quad (1)$$

and therefore  $\mathbb{E}[[M]_T] = \mathbb{E}[T] < \infty$ , by monotone convergence. It follows that  $\mathbb{E}[|N|_T] \leq \mathbb{E}[[M]_T] + \mathbb{E}[T] < \infty$  and  $\mathbb{E}[N_T] = \mathbb{E}[[M]_T - T] = 0$ , so  $N$  is a martingale. (Apply [17], Theorem I.21 to  $N$  stopped at  $t$  for any  $t \geq 0$ .) Furthermore, since  $(1_{A_n \setminus A_m} \cdot [M])_t \leq (1_{A_m^c} \cdot [M])_t$  for all  $m \leq n$  and  $t \geq 0$ , where  $A_m^c := (\mathbb{R}_+ \times \Omega) \setminus A_m$ , the sequence  $(1_{A_n \cap ([0, t] \times \Omega)} \cdot M)_{n \geq 1}$  is Cauchy in  $H^2$ , so convergent there; it follows (by [17], Theorem IV.32, say) that  $M$  stopped at  $t$  is an  $H^2$ -martingale.  $\square$

**Theorem 1.6.** If  $M$  is a normal martingale and  $T$  is a stopping time such that  $\mathbb{P}(T < \infty) > 0$  then  $N : t \mapsto 1_{T < \infty}(M_{t+T} - M_T)$  is a normal martingale (for the measure  $\tilde{\mathbb{P}} := \mathbb{P}(\cdot | T < \infty)$  and the filtration  $(\tilde{\mathcal{F}}_{t+T})_{t \geq 0}$ ).

**Proof.** As  $M$  and  $t \mapsto (M_t - M_0)^2 - t$  are local martingales, so are  $N$  and

$$\begin{aligned} Q : t \mapsto 1_{T < \infty}((M_{t+T} - M_0)^2 - (t + T) - (M_T - M_0)^2 + T) \\ = 1_{T < \infty}((M_{t+T} - M_T)^2 - t + 2(M_T - M_0)(M_{t+T} - M_T)), \end{aligned}$$

by Lemma 1.2. Hence  $t \mapsto (N_t - N_0)^2 - t = Q_t - 21_{T < \infty}(M_T - M_0)N_t$  is also a local martingale (as local martingales form a module over the algebra of random variables which are measurable with respect to the initial  $\sigma$ -algebra) and the conclusion follows from Theorem 1.5.  $\square$

**Lemma 1.7.** If  $A$  is a right-continuous, increasing process such that  $A_0 \geq 0$  and each  $A_t$  is a stopping time then  $(\mathcal{F}_{A_t})_{t \geq 0}$  is a filtration which satisfies the usual conditions.

**Proof.** This is a straightforward exercise.  $\square$

**Lemma 1.8.** *Let  $K$  and  $L$  be independent martingales and let  $A$  be a continuous, increasing,  $(\mathcal{F}_t^K)_{t \geq 0}$ -adapted process with  $A_0 = 0$  and  $A_\infty = \infty$ , where  $(\mathcal{F}_t^K)_{t \geq 0}$  denotes the smallest filtration satisfying the usual hypotheses to which  $K$  is adapted.*

*If  $\mathcal{G}_t := \mathcal{F}_t^K \vee \mathcal{F}_t^L$  for all  $t \geq 0$  then each  $A_t$  is a  $(\mathcal{G}_t)_{t \geq 0}$ -stopping time,  $(\mathcal{G}_{A_t})_{t \geq 0}$  is a filtration satisfying the usual conditions,  $L_A$  is a  $(\mathcal{G}_{A_t})_{t \geq 0}$ -local martingale and  $[L_A] = [L]_A$ . If  $H$  is an  $(\mathcal{F}_t^L)_{t \geq 0}$ -predictable process which is  $L$  integrable then  $H_A$  is  $(\mathcal{G}_{A_t})_{t \geq 0}$  predictable and  $L_A$  integrable, with  $(H \cdot L)_A = H_A \cdot L_A$ .*

*If  $\mathcal{H}_t := \mathcal{F}_t^K \vee \mathcal{F}_t^{L_A}$  for all  $t \geq 0$  then  $\mathcal{H}_t \subseteq \mathcal{G}_{A_t}$  for all  $t \geq 0$ . If there exist disjoint,  $(\mathcal{H}_t)_{t \geq 0}$ -predictable sets  $B$  and  $C$  such that  $1_B \cdot [K] = [K]$  and  $1_C \cdot [L]_A = [L]_A$  and if  $([K] + [L]_A)^{1/2}$  is  $(\mathcal{H}_t)_{t \geq 0}$ -locally integrable then  $K + L_A$  is a  $(\mathcal{H}_t)_{t \geq 0}$ -local martingale and  $[K + L_A] = [K] + [L]_A$ .*

**Proof.** This is immediate from Lemmes 1–3 and Théorème 1 of [21].  $\square$

## 2. The processes $X$ and $Y$

**Definition 2.1.** *Let  $X$  be the normal martingale which satisfies the (time-inhomogeneous) structure equation*

$$d[X]_t = (1 - t - X_{t-}) dX_t + dt$$

*with initial condition  $X_0 = 0$  and let  $Y_t := X_t + t$  for all  $t \geq 0$ . (The process  $X$  was introduced in [7], where it was constructed from the quantum stochastic analogue of the Poisson process for monotone independence. Existence also follows directly from [23], Théorème 4.0.2; uniqueness (in law) and the chaotic-representation property hold by [2], Corollary 26.) Then  $Y_0 = 0$  and*

$$d[Y]_t = (1 - Y_{t-}) dY_t + Y_{t-} dt, \quad (2)$$

*which implies that  $\Delta Y_t \in \{0, 1 - Y_{t-}\}$  for all  $t > 0$ . If*

$$G_t := \sup\{s \in [0, t] : Y_s = 1\} \in \{-\infty\} \cup ]0, t] \quad (3)$$

*then (by [7], Theorem 24)*

$$Y_t = -W_\bullet(-\exp(-1 - t + G_t)) \quad (4)$$

*for all  $t \geq 0$ , where  $W_\bullet = W_{-1}$  if  $Y_t \geq 1$  and  $W_\bullet = W_0$  if  $Y_t \leq 1$ ; a little more will be said in Proposition 6.3. (It follows from this description of the trajectories that  $X$  and  $Y$  are uniformly bounded on  $[0, t]$  for all  $t \geq 0$ .)*

**Definition 2.2.** *Let*

$$a : \mathbb{R}_+ \rightarrow ]0, 1]; \quad t \mapsto -W_0(-e^{-1-t}),$$

$$b : \mathbb{R}_+ \rightarrow [1, \infty[; \quad t \mapsto -W_{-1}(-e^{-1-t})$$

*and*

$$c : ]0, \infty[ \rightarrow \mathbb{R}_+; \quad t \mapsto b'(t) - a'(t) = \frac{b(t)}{b(t) - 1} + \frac{a(t)}{1 - a(t)}.$$

*Note that  $a(0) = b(0) = 1$ , both  $a$  and  $b$  are homeomorphisms (which may be verified by inspecting their derivatives on  $]0, \infty[$ ) and  $c(t) \searrow 1$  as  $t \rightarrow \infty$ .*

**Lemma 2.3.** For all  $t \geq 0$  the random variable  $Y_t$  is distributed with an atom at 0 (of mass  $e^{-t}$ ) and a continuous part with support  $[a(t), b(t)]$ :

$$\mathbb{P}(Y_t \in A) = \mathbb{1}_{0 \in A} e^{-t} + \frac{1}{\pi} \int_{A \cap [a(t), b(t)]} \operatorname{Im} \frac{1}{W_{-1}(-ye^{t-y})} dy \quad \forall A \in \mathcal{B}(\mathbb{R}).$$

**Proof.** See [7], Corollary 17. □

**Remark 2.4.** The (classical) Poisson process is simpler when uncompensated; similarly, it is easier to work with  $Y$  than with  $X$ . These processes are strongly Markov (by [2], Theorem 37, for example) and Émery's Itô formula ([13], Proposition 2) implies that, if  $f: \mathbb{R} \rightarrow \mathbb{R}$  is twice continuously differentiable,

$$f(X_t) = f(0) + \int_0^t g(X_{s-}, s) dX_s + \int_0^t h(X_{s-}, s) ds \quad (5)$$

and

$$f(Y_t) = f(0) + \int_0^t g(Y_{s-}, 0) dX_s + \int_0^t (h(Y_{s-}, 0) + f'(Y_{s-})) ds \quad (6)$$

for all  $t \geq 0$ , where  $g, h: \mathbb{R}^2 \rightarrow \mathbb{R}$  are such that

$$g(x, t) = \mathbb{1}_{x \neq 1-t} \frac{f(1-t) - f(x)}{1-x-t} + \mathbb{1}_{x=1-t} f'(1-t)$$

and

$$h(x, t) = \mathbb{1}_{x \neq 1-t} \frac{f(1-t) - f(x) - (1-x-t)f'(x)}{(1-x-t)^2} + \mathbb{1}_{x=1-t} \frac{1}{2} f''(1-t)$$

for all  $x, t \in \mathbb{R}$ . It follows that

$$\lim_{\varepsilon \rightarrow 0+} \frac{1}{\varepsilon} \mathbb{E}[f(X_{t+\varepsilon}) - f(X_t) | \mathcal{F}_t] = (\Gamma_t^X f)(X_t)$$

and

$$\lim_{\varepsilon \rightarrow 0+} \frac{1}{\varepsilon} \mathbb{E}[f(Y_{t+\varepsilon}) - f(Y_t) | \mathcal{F}_t] = (\Gamma_t^Y f)(Y_t),$$

for almost all  $t \geq 0$ , where

$$\begin{aligned} (\Gamma_t^X f)(x) &:= \begin{cases} \frac{f(1-t) - f(x) - (1-x-t)f'(x)}{(1-x-t)^2} & \text{if } x \neq 1-t, \\ \frac{1}{2} f''(1-t) & \text{if } x = 1-t, \end{cases} \\ &= \mathbb{1}_{x=1-t} \frac{1}{2} f''(x) + \int_{\mathbb{R} \setminus \{x\}} (f(y) - f(x) - (y-x)f'(x)) \frac{\delta_{1-t}(dy)}{(y-x)^2}, \end{aligned} \quad (7)$$

$$\begin{aligned} (\Gamma_t^Y f)(x) &:= \begin{cases} \frac{f(1) - f(x) - x(1-x)f'(x)}{(1-x)^2} & \text{if } x \neq 1, \\ \frac{1}{2} f''(1) + f'(1) & \text{if } x = 1, \end{cases} \\ &= \mathbb{1}_{x=1} \frac{1}{2} f''(x) + f'(x) + \int_{\mathbb{R} \setminus \{x\}} (f(y) - f(x) - (y-x)f'(x)) \frac{\delta_1(dy)}{(y-x)^2}, \end{aligned} \quad (8)$$

and  $\delta_z$  denotes the Dirac measure on  $\mathbb{R}$  with support  $\{z\}$ .

### 3. The final jump time

**Proposition 3.1.** *If  $G_\infty := \sup\{G_t: t \geq 0\}$ , where  $G_t$  is defined in (3), then the random variable  $G_\infty$  (the final jump time of  $Y$ ) is almost surely finite and has density*

$$g_\infty: \mathbb{R} \rightarrow \mathbb{R}_+; \quad x \mapsto \mathbb{1}_{x \geq 0} \frac{1}{\pi} \operatorname{Im} \frac{1}{W_{-1}(-e^{-1+x})}. \quad (9)$$

**Proof.** Note first that  $G_t = 1 + t - Y_t + \log Y_t$  for all  $t \geq 0$ , by (4), so  $G_t$  is  $\mathcal{F}_t$  measurable. As  $t \mapsto G_t$  is increasing, it is elementary to verify that

$$G_\infty = \sup\{G_t: t \geq 0\} = \sup\{G_n: n \geq 1\} = \lim_{n \rightarrow \infty} G_n;$$

in particular,  $G_\infty$  is  $\mathcal{F}$  measurable. If  $t > 0$  then  $\mathbb{1}_{G_n \in ]0, t]} \rightarrow \mathbb{1}_{G_\infty \in ]0, t]}$ , because  $G_n \nearrow G_\infty$ , and the dominated-convergence theorem implies that

$$\mathbb{P}(G_\infty \in ]0, t]) = \mathbb{E}[\mathbb{1}_{G_\infty \in ]0, t]}] = \lim_{n \rightarrow \infty} \mathbb{E}[\mathbb{1}_{G_n \in ]0, t]}] = \lim_{n \rightarrow \infty} \mathbb{P}(G_n \in ]0, t]).$$

Since  $\mathbb{P}(G_\infty = -\infty) = \mathbb{P}(Y \equiv 0) \leq \mathbb{P}(Y_t = 0) = e^{-t} \rightarrow 0$  as  $t \rightarrow \infty$ , it follows that  $\mathbb{P}(G_\infty = -\infty) = 0$  and

$$\mathbb{P}(G_\infty \leq t) = \lim_{n \rightarrow \infty} \mathbb{P}(G_n \in ]0, t])$$

for all  $t \geq 0$ . If  $n \geq 1$  and  $t \in [0, n]$  then

$$\begin{aligned} 0 < 1 + n - Y_n + \log Y_n \leq t &\iff -e^{-1-n} > -Y_n \exp(-Y_n) \geq -e^{-1-n+t} \\ &\iff Y_n \in ]a(n), a(n-t)] \cup [b(n-t), b(n)[ \end{aligned}$$

and, by Lemma 2.3,

$$\gamma_n(t) := \mathbb{P}(G_n \in ]0, t]) = \frac{1}{\pi} \int_{]a(n), a(n-t)] \cup [b(n-t), b(n)[} \operatorname{Im} \frac{1}{W_{-1}(-ye^{n-y})} dy.$$

Note that  $\gamma_n$  is continuously differentiable on  $[0, n[$ , with

$$\gamma'_n(s) = \frac{1}{\pi} \operatorname{Im} \frac{1}{W_{-1}(-e^{-1+s})} (b'(n-s) - a'(n-s)) = c(n-s)g_\infty(s)$$

for all  $s \in [0, n[$ . If  $n > t$  and  $s \in [0, t]$  then, by the remarks in Definition 2.2,  $\gamma'_n(s) \searrow g_\infty(s)$  as  $n \rightarrow \infty$  and the monotone-convergence theorem implies that

$$\lim_{n \rightarrow \infty} \int_0^t \gamma'_n(s) ds = \int_0^t g_\infty(s) ds \quad \forall t \geq 0.$$

This gives the result, because  $\int_0^\infty g_\infty(s) ds = 1$  (by Proposition B.1). □

**Remark 3.2.** *It follows from Proposition 3.1 that  $\mathbb{E}[G_\infty] = \infty$ ; a proof is given in Proposition B.1.*

**Remark 3.3.** *Calling  $G_\infty$  the final jump time is perhaps a little misleading, since it is not a stopping time; it is, however, almost surely the limit of a sequence of jump times. (See Corollary 6.2 and Corollary 6.4.)*

**Proposition 3.4.**  $\lim_{t \rightarrow \infty} \mathbb{P}(Y_t \leq 1) = 0$ .

**Proof.** By Lemma 2.3,

$$\mathbb{P}(Y_t \leq 1) = e^{-t} + \frac{1}{\pi} \int_{a(t)}^1 \operatorname{Im} \frac{1}{W_{-1}(-ye^{t-y})} dy \quad \forall t \geq 0. \quad (10)$$

If  $y \in ]0, 1]$  then there exists  $x \in [0, \infty[$  such that  $y = a(x)$ , and if  $t \geq x$  then

$$\operatorname{Im} \frac{1}{W_{-1}(-ye^{t-y})} = \operatorname{Im} \frac{1}{W_{-1}(-e^{-1+t-x})} = \pi g_\infty(t-x) \rightarrow 0$$

as  $t \rightarrow \infty$ . (This last claim follows from Proposition B.1.) Furthermore, as  $g_\infty$  is bounded, the integrand in (10) is bounded uniformly in  $y$  and  $t$ , so the result follows from the dominated-convergence theorem.  $\square$

**Corollary 3.5.** *As  $t \rightarrow \infty$ , the process  $Y_t \rightarrow \infty$  almost surely.*

**Proof.** If  $G_\infty < \infty$  then, as  $t \rightarrow \infty$ , either  $Y_t \rightarrow 0$  or  $Y_t \rightarrow \infty$ ; furthermore,

$$\{G_\infty < \infty\} \cap \left\{ \lim_{t \rightarrow \infty} Y_t = \infty \right\} = \{G_\infty < \infty\} \cap \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} \{Y_m > 1\}.$$

Since  $\mathbb{P}(G_\infty < \infty) = 1$  and  $\mathbb{P}(Y_n \leq 1) \rightarrow 0$  as  $n \rightarrow \infty$ , it follows that

$$\mathbb{P}\left(\lim_{t \rightarrow \infty} Y_t = \infty\right) \geq \limsup_{n \rightarrow \infty} \mathbb{P}(Y_n > 1) = 1 - \lim_{n \rightarrow \infty} \mathbb{P}(Y_n \leq 1) = 1.$$

(The inequality in the previous line holds by [8], Theorem 4.1(i).)  $\square$

#### 4. The active period

**Proposition 4.1.** *The stopping time  $S_0 := \inf\{t > 0: Y_t = 1\}$  is exponentially distributed and has mean 1.*

**Proof.** Note that  $Y_t = 0$  only if  $Y_s = 0$  for all  $s \in [0, t]$ , by (4); the claim now follows from Lemma 2.3.  $\square$

**Lemma 4.2.** *If  $Z_t := X_{t+S_0} + S_0$  for all  $t \geq 0$  then  $Z$  is a normal martingale for the filtration  $(\mathcal{F}_{t+S_0})_{t \geq 0}$  such that  $Z_0 = 1$ , which satisfies the structure equation*

$$d[Z]_t = dt + (1 - t - Z_{t-}) dZ_t \quad (11)$$

and which is independent of  $\mathcal{F}_{S_0}$ .

**Proof.** As  $Z_t = X_{t+S_0} - X_{S_0} + 1$  for all  $t \geq 0$ , Theorem 1.6 implies that  $Z$  is a normal martingale. Furthermore,

$$[Z]_t = [X]_{t+S_0} - [X]_{S_0} = \int_{S_0}^{t+S_0} (1 - r - X_{r-}) dX_r = \int_0^t (1 - s - Z_{s-}) dZ_s$$

for all  $t \geq 0$ . (The first equality is a consequence of Lemma 1.2; the last may be shown by expressing the integrals as the limit of Riemann sums, as in [17], Theorem II.21, for example.) It now follows from [2], Theorem 25, that, for all  $t \geq 0$ , the law of  $Z_t$  conditional on  $\mathcal{F}_{S_0}$  depends only on the initial value  $Z_0 = 1$  and the coefficient functions  $\alpha: s \mapsto 1 - s$  and  $\beta \equiv -1$  restricted to  $[0, t]$ , so  $Z_t$  is independent of  $\mathcal{F}_{S_0}$ .  $\square$

**Remark 4.3.** *If  $t \geq 0$  then*

$$Z_t + t = Y_{t+S_0} = -W_\bullet(-\exp(-1 - (t + S_0) + G_{t+S_0})) \in [a(t), b(t)],$$

since  $G_{t+S_0} \geq S_0$ . Consequently,  $Z$  is uniformly bounded on  $[0, t]$  for all  $t \geq 0$ .



**Remark 4.4.** Let  $m_n(t) := \mathbb{E}[(Z_t + t)^n]$  for all  $n \geq 1$  and  $t \geq 0$ , where  $Z$  is as in Lemma 4.2. It may be shown using Émery's Itô formula ([13], Proposition 2 and the subsequent remark) that

$$m_n(t) - m_{n-1}(t) = n \int_0^t m_{n-1}(s) ds \quad (12)$$

for all  $n \geq 1$  and  $t \geq 0$  (where  $m_0 \equiv 1$ ). Hence (compare [6], Section 4)

$$\hat{m}_n(p) = p^{-1} \prod_{j=1}^n (1 + jp^{-1})$$

if  $n \geq 1$ , where  $\hat{f}$  denotes the Laplace transform of  $f$ , and so

$$m_n(t) = 1 + \sum_{k=1}^n \left( \sum_{1 \leq j_1 < \dots < j_k \leq n} j_1 \cdots j_k \right) \frac{t^k}{k!} = \sum_{k=0}^n \left[ \begin{matrix} n+1 \\ n+1-k \end{matrix} \right] \frac{t^k}{k!} \quad (13)$$

for all  $t \geq 0$ , where  $\left[ \cdot \right]$  denotes the unsigned Stirling numbers of the first kind [15]. (The final identity holds by [7], Proposition 3 and Remark 6, for example.)

**Theorem 4.5.** If  $t > 0$  then  $Z_t + t = Y_{t+S_0}$  is continuously distributed, with density

$$f_{Z_t+t} : \mathbb{R} \rightarrow \mathbb{R}_+; \quad z \mapsto \mathbb{1}_{z \in [a(t), b(t)]} \frac{1}{\pi} \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{t-z})}. \quad (14)$$

**Proof.** Let  $x \geq 0$ . Since  $Y_t = \mathbb{1}_{t \geq S_0}(Z_{(t-S_0)^+} + t - S_0)$  for all  $t \geq 0$ , it follows that

$$\begin{aligned} \mathbb{P}(0 < Y_t \leq x) &= \mathbb{P}(S_0 \leq t \text{ and } Z_{(t-S_0)^+} + t - S_0 \leq x) \\ &= \int_0^t \int_{-\infty}^{x-t+s} dF_{Z_{t-s}}(z) e^{-s} ds \\ &= e^{-t} \int_0^t \int_{-\infty}^{x-u} dF_{Z_u}(z) e^u du, \end{aligned}$$

where  $F_V$  denotes the distribution function of the random variable  $V$ . (For the second equality, note that

$$\begin{aligned} \mathbb{E}[\mathbb{1}_{S_0 \leq t} \mathbb{1}_{Z_{(t-S_0)^+} + t - S_0 \leq x}] &= \mathbb{E}[\mathbb{1}_{S_0 \leq t} \mathbb{E}[\mathbb{1}_{Z_{(t-S_0)^+} + t - S_0 \leq x} | \mathcal{F}_{S_0}]] \\ &= \mathbb{E}[\mathbb{1}_{S_0 \leq t} \mathbb{E}[\mathbb{1}_{Z_{(t-S_0)^+} + t - S_0 \leq x} | S_0]], \end{aligned}$$

since  $Z$  is independent of  $\mathcal{F}_{S_0}$ .) Hence

$$\mathbb{P}(Z_t + t \leq x) = e^{-t} \frac{d}{dt} (e^t \mathbb{P}(0 < Y_t \leq x)) = \mathbb{P}(0 < Y_t \leq x) + \frac{d}{dt} \mathbb{P}(0 < Y_t \leq x).$$

Thus if  $t > 0$  then either  $x \leq a(t)$ , so that  $\mathbb{P}(Z_t + t \leq x) = 0$ , or  $x \geq b(t)$ , whence  $\mathbb{P}(Z_t + t \leq x) = 1 - e^{-t} + e^{-t} = 1$ , or  $x \in ]a(t), b(t)[$ , in which case

$$\begin{aligned} \pi \mathbb{P}(Z_t + t \leq x) &= \int_{a(t)}^x \operatorname{Im} \frac{1}{W_{-1}(-ze^{t-z})} dz - a'(t) \operatorname{Im} \frac{1}{W_{-1}(-a(t)e^{t-a(t)})} + \int_{a(t)}^x \frac{\partial}{\partial t} \operatorname{Im} \frac{1}{W_{-1}(-ze^{t-z})} dz \\ &= \int_{a(t)}^x \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{t-z})} dz, \end{aligned}$$

as claimed. (This formal working is a little awkward to justify: a rigorous proof is provided by Proposition C.2.)  $\square$

**Proposition 4.6.** *The random variables  $S_0$  and  $J := G_\infty - S_0$  are independent and  $J$  is continuous, with density*

$$f_J : \mathbb{R} \rightarrow \mathbb{R}_+; \quad x \mapsto \mathbb{1}_{x>0} \frac{1}{\pi} \operatorname{Im} \frac{1}{1 + W_{-1}(-e^{-1+x})}. \quad (15)$$

**Proof.** To see that  $S_0$  and  $J$  are independent, note first that

$$J = \lim_{n \rightarrow \infty} G_{n+S_0} - S_0 = \lim_{n \rightarrow \infty} (1 - Z_n + \log(Z_n + n))$$

almost surely, where  $(Z_t)_{t \geq 0}$  is as defined in Lemma 4.2, which implies that  $G_{n+S_0} - S_0$  is independent of  $S_0$  for all  $n \geq 1$  and, therefore, so is  $J$ .

If  $F_J(z) := \mathbb{P}(J \leq z)$  for all  $z \in \mathbb{R}$  then, by independence and Proposition 4.1,

$$\begin{aligned} \int_{-\infty}^z g_\infty(w) dw &= \mathbb{P}(J + S_0 \leq z) = \int \int_{\{(x,y) \in \mathbb{R}^2: x+y \leq z\}} dF_J(x) \mathbb{1}_{y \geq 0} e^{-y} dy \\ &= \int_{-\infty}^z e^{-v} \int_{-\infty}^v e^u dF_J(u) dv \end{aligned}$$

for all  $z \in \mathbb{R}$ , using the substitution  $(u, v) = (x, x + y)$ . Thus, for almost all  $v \in \mathbb{R}$ ,

$$g_\infty(v) = e^{-v} \int_{-\infty}^v e^u dF_J(v);$$

in fact, this holds for all  $v \in \mathbb{R}$ , as both functions are continuous, and, since  $g_\infty(0) = 0$ ,

$$g_\infty(t) = e^{-t} \int_0^t e^s dF_J(s) \quad \forall t \geq 0.$$

Now  $g_\infty$  is continuously differentiable on  $\mathbb{R} \setminus \{0\}$  and  $f_J(x) = g_\infty(x) + \mathbb{1}_{x \neq 0} g'_\infty(x)$ , so if  $0 < \varepsilon < t$  then integration by parts yields the equality

$$\int_\varepsilon^t e^s f_J(s) ds = e^t g_\infty(t) - e^\varepsilon g_\infty(\varepsilon) \rightarrow \int_0^t e^s dF_J(s) \quad \text{as } \varepsilon \rightarrow 0+.$$

Hence  $\int_0^t e^s f_J(s) ds$  exists for all  $t \geq 0$  (as does  $\int_0^t f_J(s) ds$ , by comparison) and

$$\mu : \mathcal{B}(\mathbb{R}_+) \rightarrow \mathbb{R}_+; \quad A \mapsto \int_A e^s dF_J(s) = \int_A e^s f_J(s) ds$$

is a positive Borel measure on  $\mathbb{R}_+$ ; by [19], Theorem 1.29,

$$\int_0^t f_J(s) ds = \int_0^t e^{-s} d\mu(s) = \int_0^t dF_J(s) = F_J(t) - F_J(0)$$

for all  $t \geq 0$  and

$$1 = \lim_{t \rightarrow \infty} F_J(t) = F_J(0) + \int_0^\infty g_\infty(s) ds + \lim_{t \rightarrow \infty} g_\infty(t) = F_J(0) + 1,$$

by Proposition B.1, so  $F_J(0) = 0$ . The result follows.  $\square$

**Remark 4.7.** *The distribution of  $J$  may also be found by imitating the proof of Proposition 3.1, with  $Z_t + t$  replacing  $Y_t$ , since  $J$  has the same relationship to  $Z$  as  $G_\infty$  does to  $X$ .*

**Proposition 4.8.** *If  $t \geq 0$  then*

$$\mathbb{P}(G_\infty \leq t) = -\frac{1}{\pi} \operatorname{Im} \left( W_{-1}(-e^{-1+t}) + \frac{1}{W_{-1}(-e^{-1+t})} \right) \quad (16)$$

and

$$\mathbb{P}(J \leq t) = -\frac{1}{\pi} \operatorname{Im} W_{-1}(-e^{-1+t}) = \mathbb{P}(G_\infty \leq t) + g_\infty(t). \quad (17)$$

**Proof.** These follow immediately from the identities

$$\int_0^t \frac{1}{W_{-1}(-e^{-1+x})} dx = t - \frac{(1 + W_{-1}(-e^{-1+t}))^2}{W_{-1}(-e^{-1+t})}$$

and

$$\int_0^t \frac{1}{1 + W_{-1}(-e^{-1+x})} dx = t - W_{-1}(-e^{-1+t}) - 1,$$

which are valid for all  $t \geq 0$  and may be verified by differentiation. For brevity, let  $w = W_{-1}(-e^{-1+t})$  and  $w' = W'_{-1}(-e^{-1+t})$ ; note that  $dw/dt = -e^{-1+t}w'$  and  $-e^{-1+t}(1+w)w' = w$ , whence

$$\begin{aligned} \frac{d}{dt} \left( t - \frac{(1+w)^2}{w} \right) &= 1 - \frac{-2e^{-1+t}(1+w)w'w + e^{-1+t}w'(1+w)^2}{w^2} \\ &= 1 - \frac{-e^{-1+t}(1+w)w'(2w - (1+w))}{w^2} = 1 - \frac{w-1}{w} = \frac{1}{w} \end{aligned}$$

and, if  $t > 0$ ,

$$\frac{d}{dt}(t-w) = 1 + e^{-1+t}w' = 1 - \frac{w}{1+w} = \frac{1}{1+w},$$

as required. (To see the existence of  $\int_0^t 1/(1 + W_{-1}(-e^{-1+x})) dx$ , note that if  $t \geq \varepsilon > 0$  then, letting  $W_{-1}(-e^{-1+x}) = -v \cot v + iv$ , where  $v \in ]-\pi, 0[$ ,

$$\int_\varepsilon^t \left| \frac{1}{1 + W_{-1}(-e^{-1+x})} \right| dx = \int_{v(t)}^{v(\varepsilon)} \sqrt{\frac{1 - 2v \cot v + v^2 \operatorname{cosec}^2 v}{v^2}} dv$$

and the function  $v \mapsto (1 - 2v \cot v + v^2 \operatorname{cosec}^2 v)/v^2$  is continuous on  $]-\pi, 0[$  with limit 1 as  $v \rightarrow 0-$ .)  $\square$

## 5. La martingale cachée

The martingale  $H$  discussed in this section was discovered by Émery [14].

**Theorem 5.1.** *If  $H_t := 1 - (Z_t + t)^{-1}$  for all  $t \geq 0$  then  $H$  is a martingale such that  $H_0 = 0$ ,*

$$d[H]_t = (1 - H_{t-})^2 dt - H_{t-} dH_t \quad (18)$$

and  $H_t \rightarrow H_\infty := 1$  almost surely as  $t \rightarrow \infty$ .

**Proof.** If  $t \geq 0$  and  $\mathcal{E}(-Z)$  denotes the Doléans-Dade exponential of the normal martingale  $-Z$  then  $\mathcal{E}(-Z)$  is square integrable on  $[0, t]$  for all  $t \geq 0$  and (11) implies that

$$\begin{aligned} (Z_t + t)\mathcal{E}(-Z)_t &= Z_t + t - \int_0^t (dZ_s + ds) \int_0^t \mathcal{E}(-Z)_{s-} dZ_s \\ &= Z_t + t - \int_0^t (Z_{s-} + s)\mathcal{E}(-Z)_{s-} dZ_s - \int_0^t (1 - \mathcal{E}(-Z)_{s-})(dZ_s + ds) - \int_0^t \mathcal{E}(-Z)_{s-} d[Z]_s \\ &= 1. \end{aligned}$$

Thus  $H = 1 - \mathcal{E}(-Z)$  is a martingale and  $dH_t = \mathcal{E}(-Z)_{t-} dZ_t = (1 - H_{t-}) dZ_t$ , whence

$$\begin{aligned} d[H]_t &= (1 - H_{t-})^2 d[Z]_t \\ &= (1 - H_{t-})^2 (dt + (1 - (1 - H_{t-})^{-1}) dZ_t) \\ &= (1 - H_{t-})^2 dt - H_{t-} dH_t, \end{aligned}$$

as claimed. Since  $Y_t \rightarrow \infty$  almost surely as  $t \rightarrow \infty$ , by Corollary 3.5, so does  $Z_t + t = Y_{t+S_0}$ , and the final claim follows.  $\square$

**Remark 5.2.** As  $H_t = 0$  if and only if  $Z_t + t = 1$ ,

$$\mathcal{U} := \{t \geq 0: Y_t = 1\} = \{s + S_0: Y_{s+S_0} = 1\} = \{s + S_0: H_s = 0\};$$

the structure of  $\mathcal{U}$  is determined by the zero set of  $H$ .

**Definition 5.3.** Let

$$\tau: \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}_+; \quad (t, \omega) \mapsto \tau_t(\omega) := \int_0^t (1 - H_{s-}(\omega))^2 ds$$

and note that  $\tau$  is adapted to the filtration  $(\mathcal{F}_t)_{t \geq 0}$  and has paths which are continuous, strictly increasing and bi-Lipschitzian on any compact subinterval of  $\mathbb{R}_+$ , since the derivative

$$\tau'_t = (1 - H_{t-})^2 = (Z_{t-} + t)^{-2} \in [b(t)^{-2}, a(t)^{-2}]$$

for all  $t \geq 0$ . Let

$$\tau_\infty := \int_0^\infty (1 - H_{s-})^2 ds \in ]0, \infty]$$

and extend  $\tau^{-1}$  (defined pathwise) to all of  $\mathbb{R}_+$  by letting  $\tau_s^{-1} := \infty$  for all  $s \in [\tau_\infty, \infty[$ . If  $s \geq 0$  then  $\{\tau_s^{-1} \leq t\} = \{s \leq \tau_t\} \in \mathcal{F}_t$  for all  $t \geq 0$ , so  $\tau_s^{-1}$  is an  $(\mathcal{F}_t)_{t \geq 0}$  stopping time. Thus  $(\mathcal{G}_s := \mathcal{F}_{\tau_s^{-1}})_{s \geq 0}$  is a filtration which satisfies the usual conditions, by Lemma 1.7.

**Proposition 5.4.** The process  $K = (K_s := H_{\tau_s^{-1}})_{s \geq 0}$  is a martingale for the filtration  $(\mathcal{G}_s)_{s \geq 0}$  and satisfies the equation

$$[K]_s = s \wedge \tau_\infty - \int_0^s K_{r-} dK_r \quad \forall s \geq 0. \quad (19)$$

**Proof.** Fix  $s \geq 0$ ; as  $\tau_s^{-1}$  is an  $(\mathcal{F}_t)_{t \geq 0}$  stopping time,  $H^{\tau_s^{-1}}$  is a martingale for this filtration ([18], Theorem II.77.4). Let  $(T_n)_{n \geq 1}$  be an increasing sequence of stopping times which reduces the local martingale  $H_- \cdot H$  and note that

$$\mathbb{E}[\tau_{\tau_s^{-1} \wedge T_n} - [H]_{\tau_{\tau_s^{-1} \wedge T_n}}] = \mathbb{E}[(H_- \cdot H)_{\tau_s^{-1}}^{T_n}] = 0,$$

by the optional-sampling theorem. As  $\tau$  is increasing, the monotone-convergence theorem implies that

$$\mathbb{E}[s \wedge \tau_\infty] = \lim_{n \rightarrow \infty} \mathbb{E}[\tau_{\tau_s^{-1} \wedge T_n}] = \lim_{n \rightarrow \infty} \mathbb{E}[[H]_{\tau_{\tau_s^{-1} \wedge T_n}}] = \mathbb{E}[[H^{\tau_s^{-1}}]_\infty],$$

so  $H^{\tau_s^{-1}}$  is a square-integrable martingale ([17], Corollary 4 to Theorem II.27). Hence  $K$  is a martingale, by a further application of the optional-sampling theorem: if  $0 \leq r \leq s$  then

$$\mathbb{E}[K_s | \mathcal{G}_r] = \mathbb{E}[H_\infty^{\tau_s^{-1}} | \mathcal{F}_{\tau_r^{-1}}] = H_{\tau_r^{-1}} = K_r.$$

Moreover,

$$\int_0^s K_{r-} dK_r = \int_0^{\tau_s^{-1}} K_{\tau_r-} dH_r = \int_0^{\tau_s^{-1}} H_{r-} dH_r$$

(which follows from [17], Theorem II.21, for example), so

$$[K]_s = K_s^2 - K_0^2 - 2 \int_0^s K_{r-} dK_r = H_{\tau_s^{-1}}^2 - H_0^2 - 2 \int_0^{\tau_s^{-1}} H_{r-} dH_r = [H]_{\tau_s^{-1}}$$

and this equals

$$\tau_{\tau_s^{-1}} - \int_0^{\tau_s^{-1}} H_{r-} dH_r = s \wedge \tau_\infty - \int_0^s K_{r-} dK_r.$$

□

**Theorem 5.5.** *Let  $M$  be Azéma's martingale, that is, the normal martingale such that  $M_0 = 0$  and*

$$d[M]_t = dt - M_{t-} dM_t.$$

*If  $T := \inf\{t \geq 0: M_t = 1\}$  then  $M^T$  and  $K$  are identical in law.*

**Proof.** Let  $L$  be a normal martingale which is independent of  $K$  such that  $L_0 = 1$  and

$$d[L]_t = dt - L_{t-} dL_t,$$

that is,  $L$  is an Azéma's martingale started at 1; existence of such follows from [13], Proposition 5. For all  $t \geq 0$ , let

$$P_t := \mathbb{1}_{t \in [0, \tau_\infty[} K_t + \mathbb{1}_{t \in [\tau_\infty, \infty[} L_{t-\tau_\infty} = K_t + L_{(t-\tau_\infty)^+} - 1.$$

In the notation of Lemma 1.8,  $\tau_\infty = \inf\{t \geq 0: K_t = 1\}$  is a  $(\mathcal{F}_t^K)_{t \geq 0}$ -stopping time, so  $]0, \tau_\infty]$  is  $(\mathcal{F}_t^K)_{t \geq 0}$  predictable and  $\mathbb{1}_{]0, \tau_\infty]} \cdot [K] = [K]$  (since  $K = K^{\tau_\infty}$ ) whereas  $\mathbb{1}_{]0, \tau_\infty]} \cdot [L_A] = 0$ , if  $A_t := (t - \tau_\infty)^+$ , because  $(L_A)_t^{\tau_\infty} = L_{A_t \wedge \tau_\infty} = 0$  for all  $t \geq 0$ . Since  $[K]_t = 2(t \wedge \tau_\infty) - K_t^2 \leq 2t$  and  $[L]_{A_t} = 2A_t - L_{A_t}^2 \leq 2t$ , Lemma 1.8 implies that  $P = K + L_A - 1$  is a local martingale such that  $P_0 = 0$  and

$$[P]_t = [K]_t + [L]_{A_t} = t - (K_- \cdot K)_t - (L_- \cdot L)_{A_t}.$$

However,

$$\begin{aligned} [P] &= [K] + [L_A] \\ &= K^2 - 2K_- \cdot K + L_A^2 - 1 - 2L_{A-} \cdot L_A \\ &= (K + L_A - 1)^2 + 2K + 2L_A - 2 - 2KL_A - 2K_- \cdot K - 2(L_- \cdot L)_A \end{aligned}$$

and  $KL_A = P$ , so

$$P^2 - 2P_- \cdot P = [P] = P^2 - 2K_- \cdot K - 2(L_- \cdot L)_A.$$

Thus  $[P]_t = t - (P_- \cdot P)_t$ , so  $P$  is a normal martingale, by Theorem 1.5, and, by uniqueness ([13], Proposition 6),  $P$  is equal to  $M$  in law. Since  $\tau_\infty = \inf\{t \geq 0: P_t = 1\}$ , the processes  $K = P^{\tau_\infty}$  and  $M^T$  are identical in law, as claimed.  $\square$

## 6. The level set $\mathcal{U}$

The level set

$$\mathcal{U} = \{t + S_0: H_t = 0\} = \tau^{-1}(\{s \in [0, \tau_\infty[: K_s = 0\}) + S_0,$$

where  $\tau$  is a homeomorphism between  $\mathbb{R}_+$  and  $[0, \tau_\infty[$  which is bi-Lipschitzian on compact subintervals. This observation, together with Theorem 5.5, leads immediately to the following theorem, thanks to well-known properties of the zero set of Azéma's martingale (or rather, by [17], Section IV.6, properties of the zero set of Brownian motion: see [8], Theorem 37.4 and [24]).

**Theorem 6.1.** *The set  $\mathcal{U} := \{t \geq 0: Y_t = 1\}$  is almost surely non-empty, perfect (that is, closed and without isolated points), compact and of zero Lebesgue measure. If  $a > 0$  then  $\mathcal{U} \cap [S_0, S_0 + a]$  has Hausdorff dimension  $1/2$ .*

**Corollary 6.2.** *If  $T$  is a stopping time then  $\mathbb{P}(G_\infty = T) = 0$ . In particular, the final jump time  $G_\infty$  is not a stopping time.*

**Proof.** If  $T$  is a stopping time then so is  $T' = \mathbb{1}_{Y_T=1}T + \mathbb{1}_{Y_T \neq 1}\infty$ ; let  $Z'_t := \mathbb{1}_{T' < \infty}(X_{t+T'} - X_{T'} + 1)$  for all  $t \geq 0$ . Conditional on  $T' < \infty$ , it holds that  $Z'_0 = 1$  and, working as in the proof of Lemma 4.2,

$$d[Z']_t = (1 - t - Z'_{t-})dZ'_t + dt,$$

so  $Z'$  is identical in law to  $Z$ . In particular, the set  $\mathcal{U} \cap ]T, T + 1[$  is almost surely non-empty, given that  $Y_T = 1$ , but  $\mathcal{U} \cap ]G_\infty, G_\infty + 1[ = \emptyset$  by definition.  $\square$

**Proposition 6.3.** *If  $S$  and  $T$  are random variables such that  $0 \leq S \leq T \leq \infty$  and  $Y$  is continuous on  $[S, T[$  (both almost surely) then*

$$Y_t = -W_\bullet(\exp(-1 - t + G_S)) \quad \forall t \in [S, T[$$

*almost surely, where  $\bullet \equiv 0$  or  $\bullet \equiv -1$  on  $[S, T[$ .*

**Proof.** Working pathwise, assume  $S < T$  and note that, almost surely for all  $n \geq 1$ , there exists  $T_n \in [S, S + 1/n]$  such that  $Y_{T_n} \neq 1$  (otherwise  $Y \equiv 1$  on  $[S, S + 1/n]$ , contradicting the fact that  $\mathcal{U}$  almost surely has zero Lebesgue measure). Let

$$A := \{R \in ]T_n, T]: Y \neq 1 \text{ on } [T_n, R];$$

since  $Y_{T_n} \neq 1$ , the right-continuity of  $Y$  at  $T_n$  implies that  $A$  is non-empty. Furthermore,  $R_\infty := \sup A \in A$ : there exists  $(R_n)_{n \geq 1} \subseteq A$  such that  $R_n \nearrow R_\infty$  and  $Y \neq 1$  on  $\bigcup_{n \geq 1} [T_n, R_n[ = [T_n, R_\infty[$ .

If  $R \in A$  then, working as in [7], Proof of Theorem 24, it follows that  $Y$  is continuously differentiable on  $[T_n, R[$  (taking the right derivative at  $T_n$ ) with  $Y' = Y/(Y - 1)$  there. Hence, by [7], Lemma 25,

$$Y_t = -W_\bullet(-Y_{T_n} \exp(-t + T_n - Y_{T_n})) = -W_\bullet(-\exp(-1 - t + G_{T_n}))$$

for all  $t \in [T_n, R[$ , where  $\bullet \equiv -1$  or  $\bullet \equiv 0$ . In particular,  $Y_{R_-} \neq 1$ , so if  $R_\infty < T$  then  $Y$  is continuous at  $R_\infty$  and  $Y_{R_\infty} \neq 1$ , but then there exists  $\Delta > 0$  such that  $R_\infty + \Delta < T$  and  $Y \neq 1$  on  $[R_\infty, R_\infty + \Delta[$ , contradicting the definition of  $R_\infty$ . Thus  $Y$  has the desired form on  $[T_n, T[$ ; letting  $n \rightarrow \infty$ , so that  $T_n \searrow S$ , gives the result.  $\square$

**Corollary 6.4.** *If  $T$  is a random variable such that  $Y_T = 1$  almost surely then there exists a sequence  $(T_n)_{n \geq 1}$  of random variables such that  $T_n \nearrow T$  and  $\Delta Y_{T_n} \neq 0$  almost surely.*

**Proof.** Let  $T_n := \sup\{t \in ]0, T]: |\Delta Y_t| > 1/(n+1)\}$  for all  $n \geq 1$ ; the sequence  $(T_n)_{n \geq 1}$  is increasing, with each  $T_n$  almost surely finite and such that  $\Delta Y_{T_n} \neq 0$  (since  $Y$  has càdlàg paths, so only finitely many jumps of magnitude strictly greater than  $1/(n+1)$  on any bounded interval). If  $S := \lim_{n \rightarrow \infty} T_n$  then  $Y$  is continuous on  $[S, T[$  and Proposition 6.3 implies that  $S = T$  almost surely, as required.  $\square$

## 7. Local time

This section is heavily influenced by [17], Section IV.6, hence the proofs are only sketched. Thanks to Theorem 5.5, the results may also be deduced simply from the corresponding properties of Azéma's martingale (except, perhaps, for (21)).

**Definition 7.1.** *Let  $\mathcal{P}$  denote the predictable  $\sigma$ -algebra on  $\mathbb{R}_+ \times \Omega$ . Recall (see [22], Section I.6, for example) that there exists a  $\mathcal{B}(\mathbb{R}) \otimes \mathcal{P}$ -measurable function*

$$L: \mathbb{R} \times \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}; \quad (v, t, \omega) \mapsto L_t^v(\omega)$$

such that, for all  $v \in \mathbb{R}$ ,  $L^v$  is a continuous, increasing process with  $L_0^v = 0$  and

$$\begin{aligned} |Y_t - v| &= |v| + \int_0^t \operatorname{sgn}(Y_{s-} - v) dY_s \\ &\quad + \sum_{0 < s \leq t} (|Y_s - v| - |Y_{s-} - v| - \operatorname{sgn}(Y_{s-} - v) \Delta Y_s) + L_t^v \end{aligned} \quad (20)$$

for all  $t \geq 0$  almost surely, where  $\operatorname{sgn}(x) := \mathbf{1}_{x>0} - \mathbf{1}_{x \leq 0}$  for all  $x \in \mathbb{R}$ .

**Remark 7.2.** *Since  $X$  is purely discontinuous ([7], Lemma 23),  $[Y]^c = [X]^c = 0$ ; by the occupation-density formula ([17], Corollary 2 to Theorem IV.51), there exists a null set  $N \subseteq \Omega$  such that*

$$0 = \int_0^\infty [Y]_t^c(\omega) dt = \int_{-\infty}^\infty \int_0^\infty L_t^v(\omega) dt dv \quad \forall \omega \in \Omega \setminus N,$$

and so, almost surely,  $L^v \equiv 0$  on  $\mathbb{R}_+$  for almost all  $v \in \mathbb{R}$ . The following theorem gives a more exact result.

**Theorem 7.3.** *If  $v \neq 1$  then the local time  $L^v = 0$ , whereas*

$$\mathbb{E}[L_t^1] = 2 \int_0^t g_\infty(x) dx > 0 \quad (21)$$

and the random variable  $L_t^1$  is not almost surely zero for all  $t > 0$ .

**Proof.** If  $v = 0$  then (20) implies that

$$\begin{aligned} |Y_{t+S_0}| &= - \int_0^{S_0} dY_s + \int_{S_0}^{t+S_0} dY_s + 2 \sum_{0 < s \leq t+S_0} \mathbb{1}_{Y_{s-}=0} \Delta Y_s + L_{t+S_0}^0 \\ &= -1 + Y_{t+S_0} - 1 + 2 + L_{t+S_0}^0 \end{aligned}$$

for all  $t \geq 0$ , so  $L^0 = 0$ . (The first equality uses the local character of the stochastic integral ([17], Corollary to Theorem II.18).) If  $v \notin \{0, 1\}$  then the set  $\{s > 0: Y_{s-} = Y_s = v\}$  is countable and the claim follows as it does in [17], Proof of Theorem IV.63. For the remaining case, observe that the Meyer–Tanaka–Itô formula (or just [17], Theorem IV.49) yields, for all  $t \geq 0$ , the identity

$$(Y_t - 1)^+ = \int_0^t \mathbb{1}_{Y_{s-} > 1} dY_s + \frac{1}{2} L_t^1.$$

Since

$$\mathbb{E} \left[ \int_0^t \mathbb{1}_{Y_{s-} > 1} ds \right] = \mathbb{E} \left[ \int_0^t \mathbb{1}_{Y_s > 1} ds \right] = \int_0^t \mathbb{P}(Y_s > 1) ds,$$

as  $\{s > 0: Y_{s-} \neq Y_s\}$  is countable and thus has zero Lebesgue measure, it follows that

$$\mathbb{E}[L_t^1] = 2\mathbb{E}[(Y_t - 1)^+] - 2 \int_0^t \mathbb{P}(Y_s > 1) ds.$$

For all  $t \geq 0$  and  $x \geq 0$ , let  $F_{Y_t}(x) := \mathbb{P}(Y_t \leq x)$ ; Lemma 2.3 implies that

$$\mathbb{E}[(Y_t - 1)^+] = \int_1^\infty (x - 1) dF_{Y_t}(x) = \frac{1}{\pi} \int_1^{b(t)} \operatorname{Im} \frac{x - 1}{W_{-1}(-xe^{t-x})} dx = \int_0^t b(t - y) g_\infty(y) dy,$$

using the substitution  $x = b(t - y)$ , and similarly

$$\begin{aligned} \int_0^t \mathbb{P}(Y_s > 1) ds &= \frac{1}{\pi} \int_0^t \int_1^{b(s)} \operatorname{Im} \frac{1}{W_{-1}(-xe^{s-x})} dx ds \\ &= \int_0^t \int_0^s b'(s - y) g_\infty(y) dy ds = \int_0^t (b(t - y) - 1) g_\infty(y) dy. \end{aligned}$$

Combining these calculations yields (21). □

**Definition 7.4.** A semimartingale  $R$  has locally summable jumps (or satisfies Hypothesis A, in the terminology of [17]) if

$$\sum_{0 < s \leq t} |\Delta R_s| < \infty \quad \text{almost surely } \forall t > 0.$$

**Corollary 7.5.** The martingale  $X$  does not have locally summable jumps.

**Proof.** Suppose for contradiction that  $X$  (and so  $Y$ ) has locally summable jumps. By [17], Theorem IV.56, there exists a  $\mathcal{B}(\mathbb{R}) \otimes \mathcal{P}$ -measurable function

$$\tilde{L}: \mathbb{R} \times \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}_+; \quad (v, t, \omega) \mapsto \tilde{L}_t^v(\omega)$$

such that  $(v, t) \mapsto \tilde{L}_t^v(\omega)$  is jointly right continuous in  $v$  and continuous in  $t$  for all  $\omega \in \Omega$  and, for all  $v \in \mathbb{R}$ ,  $\tilde{L}^v = L^v$ . This is, however, readily seen to contradict Theorem 7.3. □



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## Appendix A. A Poisson limit theorem

The following theorem must be well known, but a reference for it (or a version with weaker hypotheses) has proved elusive.

**Theorem A.1.** *For all  $n \geq 1$  let  $(x_{n,m})_{m=1}^n$  be a collection of independent, identically distributed random variables. If there exists  $\lambda > 0$  such that*

$$\lim_{n \rightarrow \infty} n\mathbb{E}[x_{n,1}^k] = \lambda \quad \forall k \in \mathbb{N},$$

*then  $x_{n,1} + \dots + x_{n,n}$  converges in distribution to a Poisson law with mean  $\lambda$ .*

**Proof.** If  $n \geq 1$  and  $\theta \in \mathbb{R}$  then

$$\left| \mathbb{E}[\exp(i\theta(x_{n,1} + \dots + x_{n,n}))] - \left(1 + \frac{\lambda}{n}(e^{i\theta} - 1)\right)^n \right| \leq n \left| \mathbb{E}[e^{i\theta x_{n,1}}] - 1 - \frac{\lambda(e^{i\theta} - 1)}{n} \right| \left(1 + \frac{2\lambda}{n}\right)^{n-1},$$

using the fact that  $|z^n - w^n| \leq n|z - w| \max_{1 \leq k \leq n} \{|z|^{k-1}|w|^{n-k}\}$  for all  $z, w \in \mathbb{C}$ . Furthermore, because  $|e^{i\theta} - \sum_{k=0}^{2p-1} (i\theta)^k/k!| \leq \theta^{2p}/(2p)!$  for all  $\theta \in \mathbb{R}$  and  $p \geq 1$ ,

$$\begin{aligned} & n \left| \mathbb{E}[e^{i\theta x_{n,1}}] - 1 - \frac{\lambda(e^{i\theta} - 1)}{n} \right| \\ & \leq n \left| \mathbb{E} \left[ e^{i\theta x_{n,1}} - \sum_{k=0}^{2p-1} \frac{(i\theta x_{n,1})^k}{k!} \right] \right| + \sum_{k=1}^{2p-1} \frac{|\theta|^k}{k!} |n\mathbb{E}[x_{n,1}^k] - \lambda| + \lambda \left| e^{i\theta} - \sum_{k=0}^{2p-1} \frac{(i\theta)^k}{k!} \right| \\ & \leq \frac{|\theta|^{2p}(n\mathbb{E}[x_{n,1}^{2p}] + \lambda)}{(2p)!} + \sum_{k=1}^{2p-1} \frac{|\theta|^k}{k!} |n\mathbb{E}[x_{n,1}^k] - \lambda|. \end{aligned}$$

Since  $(1 + 2\lambda/n)^{n-1} \rightarrow e^{2\lambda}$  as  $n \rightarrow \infty$ , this sequence is bounded by some constant  $C$ . Fix  $\varepsilon > 0$ , choose  $p \geq 1$  such that  $2|\theta|^{2p}\lambda/(2p)! < \varepsilon/(2C)$  and choose  $n_0$  such that

$$\frac{|\theta|^k}{k!} |n\mathbb{E}[x_{n,1}^k] - \lambda| < \frac{\varepsilon}{4pC} \quad \forall n \geq n_0, k = 1, \dots, 2p;$$

the previous working shows that

$$\left| \mathbb{E}[\exp(i\theta(x_{n,1} + \dots + x_{n,n}))] - \left(1 + \frac{\lambda}{n}(e^{i\theta} - 1)\right)^n \right| < \frac{2|\theta|^{2p}\lambda C}{(2p)!} + \frac{\varepsilon}{4p} + (2p-1)\frac{\varepsilon}{4p} < \varepsilon \quad \forall n \geq n_0.$$

Hence

$$\lim_{n \rightarrow \infty} \mathbb{E}[\exp(i\theta(x_{n,1} + \cdots + x_{n,n}))] = \lim_{n \rightarrow \infty} \left(1 + \frac{\lambda}{n}(\mathrm{e}^{i\theta} - 1)\right)^n = \exp(\lambda(\mathrm{e}^{i\theta} - 1)),$$

and the result follows from the continuity theorem for characteristic functions ([8], Theorem 26.3).  $\square$

**Remark A.2.** It follows from the working above that, if  $m \geq 1$  and  $\theta \in \mathbb{R}$ ,

$$\mathbb{E}[\mathrm{e}^{i\theta x_{n,m}}] = 1 + \left(\frac{\lambda}{n}\right)(\mathrm{e}^{i\theta} - 1) + o\left(\frac{1}{n}\right) = \mathbb{E}[\mathrm{e}^{i\theta b_n}] + o\left(\frac{1}{n}\right) \rightarrow 1$$

as  $n \rightarrow \infty$ , where  $\mathbb{P}(b_n = 0) = 1 - \lambda/n$  and  $\mathbb{P}(b_n = 1) = \lambda/n$ . Thus  $x_{n,m}$  converges to 0 in distribution, and so in probability, as  $n \rightarrow \infty$ , which explains why this result is a “law of small numbers”.

## Appendix B. The probability density function $g_\infty$

**Proposition B.1.** The function

$$g_\infty : \mathbb{R} \rightarrow \mathbb{R}_+; \quad x \mapsto \mathbf{1}_{x \geq 0} \frac{1}{\pi} \operatorname{Im} \frac{1}{W_{-1}(-\mathrm{e}^{-1+x})}$$

has a global maximum  $g_\infty(x_0) \approx 0.2306509575$  at  $x_0 \approx 0.7376612533$ , is strictly increasing on  $[0, x_0]$ , is strictly decreasing on  $[x_0, \infty[$  with  $\lim_{x \rightarrow \infty} g_\infty(x) = 0$ ,

$$\int_0^\infty g_\infty(x) dx = 1 \quad \text{and} \quad \int_0^\infty x g_\infty(x) dx = \infty.$$

**Proof.** Let  $W_{-1}(-\mathrm{e}^{-1+x}) = u(x) + iv(x)$  for all  $x \geq 0$ , where  $u(x) \in \mathbb{R}$  and  $v(x) \in ]-\pi, 0]$ . Since

$$(u + iv) \exp(u + iv) = -\exp(-1 + x) \iff \begin{cases} \mathrm{e}^u (u \cos v - v \sin v) = -\mathrm{e}^{-1+x}, \\ u \sin v + v \cos v = 0, \end{cases}$$

if  $v = 0$  then  $ue^u = -\mathrm{e}^{-1+x}$ , which has no solution for  $x > 0$ , so  $v = 0$  if and only if  $x = 0$ . Suppose henceforth that  $x > 0$ ; note that  $u = -v \cot v$ ,

$$\mathrm{e}^{-v \cot v} (-v \cos v \cot v - v \sin v) = -\mathrm{e}^{-1+x} \iff x = 1 - v \cot v + \log(v \operatorname{cosec} v)$$

and  $\pi g_\infty(x) = -v/(u^2 + v^2) > 0$ . Observe that

$$\frac{du}{dv} = -\cot v + v \operatorname{cosec}^2 v = \frac{1}{\sin^2 v} (v - \sin v \cos v) < \frac{1}{\sin^2 v} (0 - \sin 0 \cos 0) = 0,$$

because  $(d/dv)(v - \sin v \cos v) = 1 - \cos 2v > 0$ , and

$$\frac{dx}{dv} = \frac{1}{v} - 2 \cot v + v \operatorname{cosec}^2 v = \frac{1}{v} (1 - v \operatorname{cosec} v)^2 - \frac{2}{\sin v} (\cos v - 1) < 0,$$

so  $u$  is a strictly increasing function of  $x$ . As  $u(0) = -1$ ,  $u$  takes its values in  $[-1, \infty[$ ; as  $v(0) = 0$ , letting  $x = 1 - v \cot v + \log(v \operatorname{cosec} v) \rightarrow \infty$  shows that  $v \rightarrow -\pi$  (since this function of  $v$  is bounded on any proper subinterval of  $]-\pi, 0[$ ) and therefore  $u \rightarrow \infty$  as  $x \rightarrow \infty$ . (In particular,  $|g_\infty(x)| \leq 1/u^2 \rightarrow 0$  as  $x \rightarrow \infty$ .) Since  $u$  is continuous, strictly increasing and maps  $[0, \infty[$  to  $[-1, \infty[$ , there exists  $x_0$  such that  $u(x_0) = -1/2$ . Moreover,

$$g'_\infty(x) = \operatorname{Im} \frac{d}{dx} \frac{1}{W_{-1}(-\mathrm{e}^{-1+x})} = \operatorname{Im} \frac{-1}{(u + iv)(1 + u + iv)} = \frac{v(2u + 1)}{(u^2 + v^2)((1 + u)^2 + v^2)},$$

so  $g'_\infty > 0$  on  $]0, x_0[$  and  $g'_\infty < 0$  on  $]x_0, \infty[$ . (The approximate values for  $x_0$  and  $g_\infty(x_0)$  were determined with the use of Maple.)

For the integrals, the substitution  $x = v$  gives that

$$\begin{aligned} \pi \int_0^\infty g_\infty(x) dx &= \int_{-\pi}^0 \frac{\sin^2 v}{v} \left( \frac{1}{v} - 2 \cot v + v \operatorname{cosec}^2 v \right) dv \\ &= \pi + \int_{-\pi}^0 \left( \frac{\sin^2 v}{v^2} - \frac{\sin 2v}{v} \right) dv = \pi + \left[ -\frac{\sin^2 v}{v} \right]_{-\pi}^0 = \pi, \end{aligned}$$

as required. Finally, if  $\varepsilon \in ]0, \pi/2[$ ,

$$\begin{aligned} \pi \int_0^\infty x g_\infty(x) dx &= \int_{-\pi}^0 (1 - v \cot v + \log(v \operatorname{cosec} v)) \left( \frac{\sin^2 v}{v^2} - \frac{\sin 2v}{v} + 1 \right) dv \\ &\geq \int_{-\pi+\varepsilon}^{-\pi/2} -v \cot v dv \geq \frac{\pi}{2} \int_\varepsilon^{\pi/2} \cot w dw = -\log \sin \varepsilon \rightarrow \infty \end{aligned}$$

as  $\varepsilon \rightarrow 0+$ . □

**Remark B.2.** It follows from Propositions B.1 and 3.1 that the distribution of  $G_\infty$  is unimodal with mode  $x_0$ , that is,  $t \mapsto \mathbb{P}(G_\infty \leq t)$  is convex on  $]-\infty, x_0[$  and concave on  $]x_0, \infty[$ .

## Appendix C. An auxiliary calculation

**Lemma C.1.** If  $f_J$  is as defined in Proposition 4.6 then

$$\pi f_J(t) = \operatorname{Im} \frac{1}{1 + W_{-1}(-e^{-1+t})} \sim \frac{1}{\sqrt{2t}} \quad \text{as } t \rightarrow 0+$$

and  $f_J$  is strictly decreasing on  $]0, \infty[$ .

**Proof.** For all  $t \geq 0$ , let  $p := -\sqrt{2(1 - e^t)} = -i\sqrt{2t} + O(t^{3/2})$  as  $t \rightarrow 0+$ ; recall that

$$-W_{-1}(-e^{-1+t}) = 1 - p + O(p^2) = 1 + i\sqrt{2t} + O(t)$$

as  $t \rightarrow 0+$ , by [10], (4.22), and this gives the first result. For the next claim, if  $t > 0$  and  $W_{-1}(-e^{-1+t}) = -v \cot v + iv$ , where  $v \in ]-\pi, 0[$ , then

$$\pi f'_J(t) = \operatorname{Im} \frac{-W_{-1}(-e^{-1+t})}{(1 + W_{-1}(-e^{-1+t}))^3} = \frac{((3 - 2v \cot v)v^2 \operatorname{cosec}^2 v - 1)v}{((1 - v \cot v)^2 + v^2)^3}.$$

The result follows if

$$(3 - 2v \cot v)v^2 \operatorname{cosec}^2 v - 1 > 0 \quad \Longleftrightarrow \quad (v^2 - \sin^2 v) \sin v + 2v^2(\sin v - v \cos v) < 0$$

for all  $v \in ]-\pi, 0[$ , but since  $\sin^2 v < v^2$  and  $\sin v < v \cos v$  for such  $v$ , this is clear. □

**Proposition C.2.** If  $D := \{(t, x) \in \mathbb{R}_+^2 : a(t) \leq x \leq b(t)\}$ ,

$$f : D \rightarrow \mathbb{R}_+; \quad (t, x) \mapsto \operatorname{Im} \frac{1}{W_{-1}(-xe^{t-x})},$$

$$F : D \rightarrow \mathbb{R}_+; \quad (t, x) \mapsto \int_{a(t)}^x f(t, y) dy$$

and  $(s, y) \in D^\circ := \{(t, x) \in \mathbb{R}_+^2 : t > 0, a(t) < x < b(t)\}$  then

$$F(s, y) + \frac{\partial F}{\partial t}(s, y) = \int_{a(s)}^y \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{s-z})} dz. \quad (22)$$

**Proof.** Note first that, since  $f$  is continuous,  $F$  is well defined. If  $h > 0$  then

$$\frac{F(s+h, y) - F(s, y)}{h} = \frac{1}{h} \int_{a(s+h)}^{a(s)} f(s+h, z) dz + \int_{a(s)}^y \frac{f(s+h, z) - f(s, z)}{h} dz$$

and the intermediate-value theorem gives  $\zeta_h \in [a(s+h), a(s)]$  such that

$$\frac{1}{h} \int_{a(s+h)}^{a(s)} f(s+h, z) dz = \frac{a(s) - a(s+h)}{h} f(s+h, \zeta_h) \rightarrow -a'(s) f(s, a(s)) = 0$$

as  $h \rightarrow 0+$ . For all  $z \in [a(s), b(s)]$  there exists  $\theta_{h,z} \in ]0, 1[$  such that

$$\frac{f(s+h, z) - f(s, z)}{h} = \frac{\partial f}{\partial t}(s + \theta_{h,z}h, z)$$

by the mean-value theorem, since  $t \mapsto f(t, z)$  is continuous on  $[s, s+h]$  and differentiable on  $]s, s+h[$ . Let

$$g: D^\circ \rightarrow \mathbb{R}_+; \quad (t, x) \mapsto \frac{\partial f}{\partial t}(t, x) + f(t, x) = \begin{cases} \pi f_J(t - a^{-1}(x)) & \text{if } x \in ]a(t), 1], \\ \pi f_J(t - b^{-1}(x)) & \text{if } x \in [1, b(t)[, \end{cases}$$

where  $f_J$  is defined in Proposition 4.6. The continuity of  $f$  on  $[s, s+1] \times [a(s), y]$  and the dominated-convergence theorem imply that

$$F(s, y) = \int_{a(s)}^y f(s, z) dz = \lim_{h \rightarrow 0+} \int_{a(s)}^y f(s + \theta_{h,z}h, z) dz,$$

so the right-hand limit in (22) has the correct value if  $\int_{a(s)}^y g(s, z) dz$  exists and

$$\lim_{h \rightarrow 0+} \int_{a(s)}^y g(s + \theta_{h,z}h, z) dz = \int_{a(s)}^y g(s, z) dz.$$

Fix  $r \in ]0, s[$  such that  $y > a(r)$  and note that  $g$  is continuous on  $[s, s+1] \times [a(r), y]$ , so the dominated-convergence theorem implies that

$$\lim_{h \rightarrow 0+} \int_{a(r)}^y g(s + \theta_{h,z}h, z) dz = \int_{a(r)}^y g(s, z) dz.$$

Next, note that if  $z \in ]a(s), a(r)]$  and  $h \rightarrow 0+$  then

$$g(s + \theta_{h,z}h, z) = \pi f_J(s + \theta_{h,z}h - a^{-1}(z)) \nearrow \pi f_J(s - a^{-1}(z)) = g(s, z),$$

because  $f_J$  is strictly decreasing, by Lemma C.1. The first half of the result now follows from the monotone-convergence theorem, once it is known that  $\int_{a(s)}^{a(r)} g(s, z) dz$  exists. However,

$$\int_{a(s)}^{a(r)} g(s, z) dz = \pi \int_s^r f_J(s - u) a'(u) du = -\pi \int_0^{s-r} f_J(t) a'(s - t) dt < \infty,$$

since, by Lemma C.1,  $\pi f_J(t) \sim 1/\sqrt{2t}$  as  $t \rightarrow 0+$ ,  $f_J$  is continuous on  $]0, s-r]$  and  $a'$  is continuous on  $[r, s]$ .

Now suppose that  $h < 0$  is such that  $s + h > 0$  and  $b(s + h) > y > a(s + h)$ . Then

$$\frac{F(s + h, y) - F(s, y)}{h} = \int_{a(s+h)}^y \frac{f(s + h, z) - f(s, z)}{h} dz - \frac{1}{h} \int_{a(s)}^{a(s+h)} f(s, z) dz$$

and the second term tends to 0 as  $h \rightarrow 0-$ . If  $z \in [a(s + h), b(s + h)]$  then  $t \mapsto f(t, z)$  is continuous on  $[s + h, s]$  and differentiable on  $]s + h, s[$ , so there exists  $\theta_{h,z} \in ]0, 1[$  such that

$$\frac{f(s + h, z) - f(s, z)}{h} = \frac{\partial f}{\partial t}(s + \theta_{h,z}h, z).$$

Furthermore, as  $f$  is continuous, so bounded, on the compact set  $D \cap ([0, s] \times \mathbb{R}_+)$ , the dominated-convergence theorem implies that

$$F(s, y) = \lim_{h \rightarrow 0-} \int_{a(s+h)}^y f(s + \theta_{h,z}h, z) dz$$

and the result follows if

$$\lim_{h \rightarrow 0-} \int_{a(s+h)}^y g(s + \theta_{h,z}h, z) dz = \int_{a(s)}^y g(s, z) dz.$$

Fix  $0 < r_1 < r_2 < s$  such that  $a(r_1) < y$  and note that  $g$  is continuous on  $[r_2, s] \times [a(r_1), y]$ , so bounded there, and the dominated-convergence theorem implies that

$$\int_{a(r_1)}^y g(s + \theta_{h,z}h, z) dz \rightarrow \int_{a(r_1)}^y g(s, z) dz$$

as  $h \rightarrow 0-$ . A final application of the monotone-convergence theorem completes the result, since if  $h < 0$  is such that  $r_2 < s + h$  then, letting  $h \rightarrow 0-$ ,

$$\begin{aligned} \mathbb{1}_{z \in [a(s+h), a(r_1)]} g(s + \theta_{h,z}h, z) &= \mathbb{1}_{z \in [a(s+h), a(r_1)]} \pi f_J(s + \theta_{h,z}h - a^{-1}(z)) \\ &\nearrow \mathbb{1}_{z \in ]a(s), a(r_1)]} \pi f_J(s - a^{-1}(z)) \\ &= \mathbb{1}_{z \in ]a(s), a(r_1)]} g(s, z). \end{aligned}$$

□

## Appendix D. A pair of Laplace transforms

**Theorem D.1.** *If  $g_\infty$  is as defined in Proposition 3.1 and  $f_J$  is as defined in Proposition 4.6 then their Laplace transforms are as follows:*

$$\widehat{g_\infty}(p) = \frac{e^{-p} p^p}{\Gamma(p+2)} \quad \text{and} \quad \widehat{f_J}(p) = (p+1) \widehat{g_\infty}(p) = \frac{e^{-p} p^p}{\Gamma(p+1)}, \quad (23)$$

where  $\Gamma: p \mapsto \int_0^\infty z^{p-1} e^{-z} dz$  is the gamma function.

**Proof.** Let

$$f_1(t) := \frac{1}{\pi} \int_{a(t)}^{b(t)} \operatorname{Im} \frac{1}{W_{-1}(-ye^{t-y})} dy \quad \forall t \geq 0.$$

Splitting the interval  $[a(t), b(t)]$  at 1 and using the substitutions  $y = a(t - x)$  and  $y = b(t - x)$ , as appropriate,

$$f_1(t) = \frac{1}{\pi} \int_0^t \operatorname{Im} \left( \frac{1}{W_{-1}(-e^{-1+x})} \right) c(t - x) dx = (g_\infty \star c)(t),$$

where  $\star$  denotes convolution of functions on  $\mathbb{R}_+$  and  $c$  is as in Definition 2.2. Furthermore,

$$\begin{aligned}\widehat{c}(p) &:= \int_0^\infty c(x)e^{-px} dx = \int_0^\infty b'(x)e^{-px} dx - \int_0^\infty a'(x)e^{-px} dx \\ &= \int_1^\infty e^{-p(-1+y-\log y)} dy + \int_0^1 e^{-p(-1+y-\log y)} dy \\ &= e^p \int_0^\infty \left(\frac{z}{p}\right)^p e^{-z} p^{-1} dz.\end{aligned}$$

The second line follows from the substitutions  $y = b(x)$  and  $y = a(x)$ . Thus, since  $f_1(t) = \mathbb{P}(Y_t > 0) = 1 - e^{-t}$ ,

$$\widehat{g_\infty}(p) = \frac{\widehat{f_1}(p)}{\widehat{c}(p)} = \frac{1}{p(p+1)} \frac{e^{-p} p^{p+1}}{\Gamma(p+1)} = \frac{e^{-p} p^p}{\Gamma(p+2)},$$

as claimed. If

$$f_2(t) := \frac{1}{\pi} \int_{a(t)}^{b(t)} \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{t-z})} dy \quad \forall t > 0,$$

then, working as above,  $f_2 = f_J \star c$ . Moreover, since  $f_2 = f_1 + f'_1$  (by the working in the proof of Theorem 4.5), it follows that  $\widehat{f_2}(p) = (p+1)\widehat{f_1}(p)$  and

$$\widehat{f_J}(p) = \frac{\widehat{f_2}(p)}{\widehat{c}(p)} = \frac{(p+1)\widehat{f_1}(p)}{\widehat{c}(p)} = \frac{e^{-p} p^p}{\Gamma(p+1)}.$$

□

**Remark D.2.** The substitution  $x = 1 - v \cot v + \log(v \operatorname{cosec} v)$  yields the identity

$$e^p \widehat{f_J}(p) = \frac{1}{\pi} \int_0^\pi \left(\frac{\sin v}{v}\right)^p \exp(pv \cot v) dv; \quad (24)$$

it should be possible to verify directly that the right-hand side of (24) equals  $p^p/\Gamma(p+1)$ . (This would give independent proof that

$$A \mapsto \mathbf{1}_{0 \in A} + \frac{1}{\pi} \int_{A \cap [a(t), b(t)]} \operatorname{Im} \frac{1}{W_{-1}(-ye^{t-y})} dy$$

and

$$A \mapsto \frac{1}{\pi} \int_{A \cap [a(t), b(t)]} \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{t-z})} dz$$

are probability measures on  $\mathcal{B}(\mathbb{R})$ .)

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